

Mixed-Integer PDE-Constrained Optimization

Applied Mathematics Research for Exascale Computing

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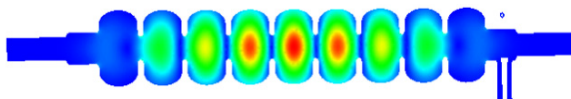
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Mixed-Integer PDE-Constrained Optimization

A new modeling paradigm

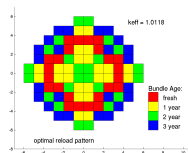
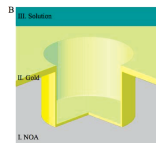
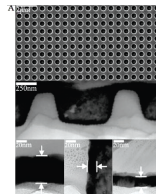
- Paradigm shift from forward simulations to **design of complex structures**
- Design of systems involving
 - ... complex PDE simulations,
 - ... uncertainty quantification,
 - ... and discrete/continuous design parameters

Integrate maths, algorithms, and CS



Shape optimization of cavity for ILC

[Akcelik et al., 2005]



Applications of MIPDECO

A rich set of DOE application areas

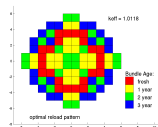
① Subsurface Design Applications

E.g. remediation of contaminated sites, oil and gas extraction

- PDE constraints model subsurface flow
- Discrete design parameters model location/operation of wells
- Uncertainties model the unknown subsurface

② Operational planning for nuclear reactors: core-reloading

- Neutron transport & fluid-flow equations
- Discrete variables model fuel rod arrangement
- Imperfect knowledge of fuel \Rightarrow uncertainties



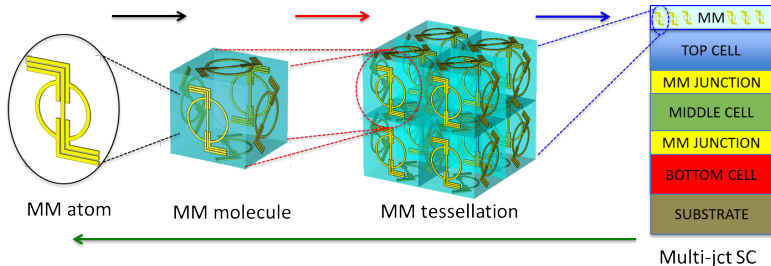
③ Accelerator design: Maxwell's equation plus discrete components, e.g. # arc & wiggler cells

④ Design of nano-materials for ultra-efficient solar cells ... next

MIPDECO for Design of Solar Cells

Goal: Design nonreciprocal coating: full transmission & absorption

- Design meta-material (MM) coating for solar cells
- MM atom of given shape is assembled into molecule
- Typical crystal (layer of molecules) has 10-20 molecules width



- Maxwell's equation models electromagnetic response
- 0 – 1 variables model orientation of atom on faces of molecule

MIPDECO for Design of Solar Cells

General form of **Maxwell's equation**

$$\nabla \times \mathbf{H} = \frac{\partial \mathbf{D}}{\partial t} + \mathbf{J}_e, \quad \nabla \times \mathbf{E} = -\frac{\partial \mathbf{B}}{\partial t} + \mathbf{J}_m,$$
$$\nabla \cdot \mathbf{D} = \rho, \quad \nabla \cdot \mathbf{B} = 0,$$

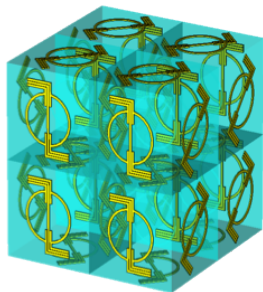
$$\begin{bmatrix} \mathbf{D} \\ \mathbf{B} \end{bmatrix} = \begin{bmatrix} \epsilon & \chi \\ \zeta & \mu \end{bmatrix} \begin{bmatrix} \mathbf{E} \\ \mathbf{H} \end{bmatrix} + \begin{bmatrix} \mathbf{P} \\ \mathbf{M} \end{bmatrix}$$

where $\epsilon = \epsilon(x)$ permittivity; μ permeability

Binary variables $z_{ijk} \in \{0, 1\}$ model orientation of MM atoms
 \Rightarrow construct permittivity and permeability tensors $\epsilon(x), \mu(x)$

$$\epsilon(x) \simeq \widetilde{\epsilon}_{j,k} = \sum_{i \in \mathcal{O}} z_{i,j,k} \epsilon_i \quad \text{and} \quad \sum_{i \in \mathcal{O}} z_{i,j,k} = 1$$

where ϵ_i is fixed permittivity of orientation i



Mathematical and Computational Challenges

Applications give rise to **MIPDECO** under **Uncertainty**

$$\begin{aligned} & \underset{y(\gamma), u, z}{\text{minimize}} && F(y(\gamma), u, z) \\ & \text{subject to} && g(y(\gamma), u, z; \gamma) = 0, \forall \gamma \in \Gamma \\ & && y(\gamma) \in \mathcal{Y}, u \in \mathcal{U}, z \in \mathbb{Z}^p \cap \mathcal{S}, \end{aligned}$$

where

- $y(\gamma)$ state variables depending on **random variables** $\gamma \in \Gamma$
- u continuous design variables
- z **integer design variables**
- g describes PDE and boundary conditions
- $F(y(\gamma), u, z)$ is objective, e.g. maximize power of solar cells



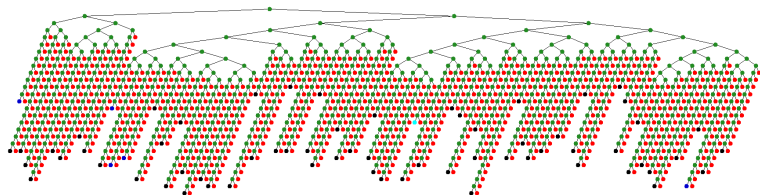
Challenge I: Combinatorial Explosion

MIPDECOs generate huge search trees [Belotti et al., 2013]

- Each node in tree is PDE-constrained optimization
- Must take uncertainty into account
- $\{4 \text{ angles}\} \times \{6 \text{ faces}\} \times \{10\text{k cubes}\} = 24\text{k binary variables}$

Brute-Force Approach: assume billion PDECOs per second

⇒ exascale machine would run longer than age of universe!



Must exploit hot-starts for re-solve ... solve millions of (N)LPs

Challenge II: New Algorithms & Math

A Simple Approach

$$\left\{ \begin{array}{l} \text{PICO [Eckstein et al., 2001]} \\ \text{MINOTAUR [Mahajan et al., 2011]} \end{array} \right\} + \left\{ \begin{array}{l} \text{PETSc} \\ \text{Trilinos} \end{array} \right\} \Rightarrow \text{failure}$$

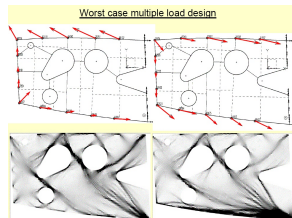
Need integrated algorithmic strategy, e.g.

- Tree-search techniques based on surrogate models
- Integrate multilevel combinatorial with multilevel PDE
- Optimization framework guides exploration of uncertainty
- Align algorithmic hierarchies with machine/storage hierarchies

... concerted effort worthy of DOE labs

New math challenges

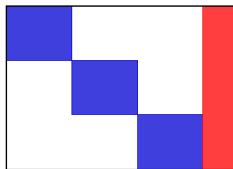
- Quality or bounds on surrogate models
- What is $\{0, 1\}^\infty$? E.g. function space for topology optimization



Integrated Uncertainty Quantification

Sources of Uncertainty

- Errors in material properties (ϵ, μ)
- Manufacturing inaccuracies
- Numerical/modeling errors
- Errors from data measurements



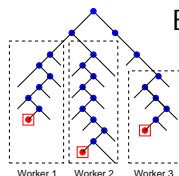
Combine PDECO with stochastic programming for UQ

- Design under UQ as **two-stage stochastic MIP optimization**
- **First-stage variables are design variables**
- **Second-stage variables are solutions of stochastic PDE**

⇒ Block-angular structure, where each block is PDE

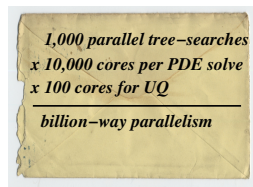
Goal: control uncertainty; dimension reduction; adjoint technology

Toward Billion-Way Concurrency



Back-of-Envelope Computation

- Asynchronous tree-search
- Linear solvers inside PDEs
- UQ block-structure



Opportunities for Exascale

- Tree-search loosely coupled solves with small communication
 - Readily scales to 1,000 parallel tree-searches [Goux and Leyffer, 2003]
 - Communicate bounds, new solutions, and sub-trees
- Parallel solvers for PDE-constrained optimization
 - Scalable linear algebra ... up to 150k cores (NEK5000)
- Scenario-based decomposition or UQ
 - Scales with number of scenarios (samples)
 - Some communication between scenarios

Hierarchy of concurrency \Rightarrow multiplicative opportunities



Algorithm-Level Resiliency

Exascale systems likely to have shorter mean-time-to-failure
... check-point-restart no longer an option

Optimization Algorithms can be made resilient

- **MIP tree-search:** only check-point master node
- Trust-region or line-search provide algorithmic resilience
- **Exploit multi-level hierarchy for smart check-pointing**
- **Stochastic programming (UQ)** robust to node failures
- Resilient linear solves [Bridges et al., 2012]

... exploit resilient algorithm design at all levels of hierarchy

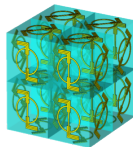
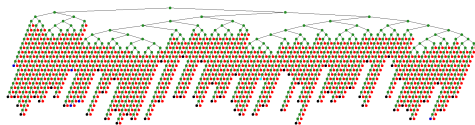


Summary and Discussion

Mixed-Integer PDE-Constrained Optimization

- Design of **complex systems** with **discrete parameters** **under uncertainty**
- Opportunity to tackle new, broader class of design applications
- Poses rich set of mathematical, algorithmic, and CS challenges
- Math and DOE applications “made for exascale”
 - Unlikely to tackle problems on smaller systems
 - Hierarchy of concurrency maps well to exascale systems
 - Integrates & combines existing DOE tools

... requires concerted & coordinated effort





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